

Math 164: Problem Set 5

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$$\begin{aligned}
 \mathbf{6.29} \quad (a) \quad & \frac{1}{n} \sum_{i=1}^n (ax_i + b - y_i)^2 \\
 &= \frac{1}{n} \sum_{i=1}^n a^2 x_i^2 + b^2 + y_i^2 - 2ax_i y_i + 2ax_i b - 2by_i \\
 &= a^2 \overline{X^2} + b^2 + \overline{Y^2} + 2a\overline{XY} + 2ab\overline{X} - 2b\overline{Y} \\
 &= \mathbf{z}^\top \begin{bmatrix} \overline{X^2} & \overline{X} \\ \overline{X} & 1 \end{bmatrix} \mathbf{z} - 2 \begin{bmatrix} \overline{XY} & \overline{Y} \end{bmatrix} \mathbf{z} + \overline{Y^2}
 \end{aligned}$$

$$(b) \text{ By the FONC } \nabla f = 2 \begin{bmatrix} \overline{X^2} & \overline{X} \\ \overline{X} & 1 \end{bmatrix} \mathbf{z} - \begin{bmatrix} \overline{XY} \\ \overline{Y} \end{bmatrix} = 0$$

$$\Rightarrow \mathbf{z}^* = \begin{bmatrix} \overline{X^2} & \overline{X} \\ \overline{X} & 1 \end{bmatrix}^{-1} \begin{bmatrix} \overline{XY} \\ \overline{Y} \end{bmatrix} \text{ is the only solution.}$$

$$= \frac{1}{\overline{X^2} - \overline{X}^2} \begin{bmatrix} 1 & -\overline{X} \\ -\overline{X} & \overline{X^2} \end{bmatrix} \begin{bmatrix} \overline{XY} \\ \overline{Y} \end{bmatrix}$$

$$= \begin{bmatrix} \frac{\overline{XY} - \overline{X} \cdot \overline{Y}}{\overline{X^2} - \overline{X}^2} \\ \frac{\overline{X^2} \cdot \overline{Y} - \overline{XY} \cdot \overline{X}}{\overline{X^2} - \overline{X}^2} \end{bmatrix}$$

$$\begin{aligned}
 (c) \text{ WTS } \overline{Y} &= \overline{X} \frac{\overline{XY} - \overline{X} \cdot \overline{Y}}{\overline{X^2} - \overline{X}^2} + \frac{\overline{X^2} \cdot \overline{Y} - \overline{XY} \cdot \overline{X}}{\overline{X^2} - \overline{X}^2} \\
 &= \frac{\overline{X} \cdot \overline{XY} - \overline{X}^2 \cdot \overline{Y}}{\overline{X^2} - \overline{X}^2} + \frac{\overline{X^2} \cdot \overline{Y} - \overline{XY} \cdot \overline{X}}{\overline{X^2} - \overline{X}^2} \\
 &= \frac{\overline{Y}(\overline{X^2} - \overline{X}^2)}{\overline{X^2} - \overline{X}^2} \\
 &= \overline{Y}
 \end{aligned}$$

$$\mathbf{6.30} \text{ Let } f(\mathbf{x}) = \frac{1}{p} \sum_{i=1}^p \|\mathbf{x} - \mathbf{x}^{(p)}\|^2 = \frac{1}{p} \sum_{i=1}^p (\mathbf{x} - \mathbf{x}^{(p)})^\top (\mathbf{x} - \mathbf{x}^{(p)})$$

$$\Rightarrow \nabla f = \frac{1}{p} \sum_{i=1}^p 2(\mathbf{x} - \mathbf{x}^{(i)}) = 0$$

$$\Rightarrow \arg \min_{\mathbf{x}} f(\mathbf{x}) = \mathbf{x}^* = \frac{1}{p} \sum_{i=1}^p \mathbf{x}^{(i)} \text{ which is just the mean (or centroid).}$$

Because the hessian $F(\mathbf{x}) = \mathbf{I}_n > 0$, the mean of f is a strict local minimizer.

- 6.31** Because Ω is convex, let $\phi(\alpha) = f(\mathbf{x}^* + \alpha \mathbf{d})$ for some feasible direction \mathbf{d} and $0 < \alpha_0$ s.t $\mathbf{x}^* + \alpha \mathbf{d} \in \Omega, \forall \alpha \in [0, \alpha_0]$. By the MVT we have $\frac{\phi(\alpha) - \phi(0)}{\alpha} = \mathbf{d}^\top \nabla f(\xi)$ for $0 \leq \xi \leq \alpha$. Because $\lim_{\alpha \rightarrow 0} \frac{\phi(\alpha) - \phi(0)}{\alpha} = \mathbf{d}^\top \nabla f(\mathbf{x}^*) \geq c \|\mathbf{d}\| > 0$, we can find an α_0 small enough s.t $\frac{\phi(\alpha) - \phi(0)}{\alpha} > 0 \Rightarrow \phi(\alpha) > \phi(0)$ for sufficiently small α . Hence, $f(\mathbf{x}^*)$ is strict local minimizer.
- 6.32** Because Ω is convex, let $\phi(\alpha) = f(\mathbf{x}^* + \alpha \mathbf{d})$ for some feasible direction \mathbf{d} and $0 < \alpha_0$ s.t $\mathbf{x}^* + \alpha \mathbf{d} \in \Omega, \forall \alpha \in [0, \alpha_0]$. By Taylor's Theorem, $\phi(\alpha) = f(\mathbf{x}^*) + \alpha \mathbf{d}^\top \nabla f(\mathbf{x}^*) + \frac{\alpha^2}{2} \mathbf{d}^\top F(\mathbf{x}^*) \mathbf{d} + o(\alpha^3)$
 $\geq f(\mathbf{x}^*) + \alpha \cdot 0 + \frac{\alpha^2}{2} c \|\mathbf{d}\|^2 + o(\alpha^3) > f(\mathbf{x}^*)$ for sufficiently small α . Hence, $f(\mathbf{x}^*)$ is strict local minimizer.
- 6.34** Let $\mathbf{u} = [u_1, u_2, \dots, u_n]^\top$. Using the system
 $x_n = \alpha x_{n-1} + \beta u_n$
 $= \alpha(\alpha x_{n-2} + \beta u_{n-1}) + \beta u_n$
 \dots
 $= \beta \sum_{i=1}^n \alpha^{n-i} u_i = \mathbf{v}^\top \mathbf{u}$ where $\mathbf{v} = [\beta \alpha^{n-1}, \beta \alpha^{n-2}, \dots, \beta]^\top$
This gives us the quadratic form minimization problem $f(\mathbf{u}) = r \mathbf{u}^\top \mathbf{u} - q \mathbf{v}^\top \mathbf{u}$ with minimizer $\mathbf{u}^* = \frac{q}{2r} \mathbf{v}$ satisfying $\nabla f(\mathbf{u}^*) = 2r \mathbf{u}^* - q \mathbf{v} = 0$ and $F(\mathbf{u}^*) = 2r \mathbf{I}_n > 0$.